UNITED STATES PATENT AND TRADEMARK OFFICE

APPLICATION NO.	FILING DATE	FIRST NAMED INVENTOR	ATTORNEY DOCKET NO.	CONFIRMATION NO.
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MICROSOFT CORPORATION
ONE MICROSOFT WAY
REDMOND, WA 98052-6399

EXAMINER
JOHNSON, JOHNISE T

ART UNIT	PAPER NUMBER
2166	

NOTIFICATION DATE	DELIVERY MODE
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Please find below and/or attached an Office communication concerning this application or proceeding.

The time period for reply, if any, is set in the attached communication.

Notice of the Office communication was sent electronically on above-indicated "Notification Date" to the following e-mail address(es):

roks@microsoft.com
ntovar@microsoft.com

	Application No.	**Applicant(s)**
Office Action Summary	10/825,488	GANTI ET AL.
	Examiner	**Art Unit**	
	Johnese Johnson	2166	

-- The MAILING DATE of this communication appears on the cover sheet with the correspondence address --

Period for Reply

A SHORTENED STATUTORY PERIOD FOR REPLY IS SET TO EXPIRE _3_ MONTH(S) OR THIRTY (30) DAYS,
WHICHEVER IS LONGER, FROM THE MAILING DATE OF THIS COMMUNICATION.
- Extensions of time may be available under the provisions of 37 CFR 1.136(a). In no event, however, may a reply be timely filed
  after SIX (6) MONTHS from the mailing date of this communication.
- If NO period for reply is specified above, the maximum statutory period will apply and will expire SIX (6) MONTHS from the mailing date of this communication.
- Failure to reply within the set or extended period for reply will, by statute, cause the application to become ABANDONED (35 U.S.C. § 133).
  Any reply received by the Office later than three months after the mailing date of this communication, even if timely filed, may reduce any
  earned patent term adjustment. See 37 CFR 1.704(b).

Status

1)☒ Responsive to communication(s) filed on _13 December 2007_.

2a)☐ This action is **FINAL**.　　2b)☒ This action is non-final.

3)☐ Since this application is in condition for allowance except for formal matters, prosecution as to the merits is
　　　closed in accordance with the practice under *Ex parte Quayle*, 1935 C.D. 11, 453 O.G. 213.

Disposition of Claims

4)☒ Claim(s) _1-21 and 25-34_ is/are pending in the application.

　　4a) Of the above claim(s) _____ is/are withdrawn from consideration.

5)☐ Claim(s) _____ is/are allowed.

6)☒ Claim(s) _1-21 and 25-34_ is/are rejected.

7)☐ Claim(s) _____ is/are objected to.

8)☐ Claim(s) _____ are subject to restriction and/or election requirement.

Application Papers

9)☐ The specification is objected to by the Examiner.

10)☐ The drawing(s) filed on _____ is/are: a)☐ accepted or b)☐ objected to by the Examiner.

　　　Applicant may not request that any objection to the drawing(s) be held in abeyance. See 37 CFR 1.85(a).

　　　Replacement drawing sheet(s) including the correction is required if the drawing(s) is objected to. See 37 CFR 1.121(d).

11)☐ The oath or declaration is objected to by the Examiner. Note the attached Office Action or form PTO-152.

Priority under 35 U.S.C. § 119

12)☐ Acknowledgment is made of a claim for foreign priority under 35 U.S.C. § 119(a)-(d) or (f).

　　a)☐ All　b)☐ Some * c)☐ None of:

　　　1.☐ Certified copies of the priority documents have been received.

　　　2.☐ Certified copies of the priority documents have been received in Application No. _____.

　　　3.☐ Copies of the certified copies of the priority documents have been received in this National Stage
　　　　　application from the International Bureau (PCT Rule 17.2(a)).

　　* See the attached detailed Office action for a list of the certified copies not received.

Attachment(s)

1)☐ Notice of References Cited (PTO-892)

2)☐ Notice of Draftsperson's Patent Drawing Review (PTO-948)

3)☐ Information Disclosure Statement(s) (PTO/SB/08)
　　Paper No(s)/Mail Date _____.

4)☐ Interview Summary (PTO-413)
　　Paper No(s)/Mail Date. _____.

5)☐ Notice of Informal Patent Application

6)☐ Other: _____.

DETAILED ACTION

Continued Examination Under 37 CFR 1.114

1.   A request for continued examination under 37 CFR 1.114, including the fee set

forth in 37 CFR 1.17(e), was filed in this application after final rejection. Since this

application is eligible for continued examination under 37 CFR 1.114, and the fee set

forth in 37 CFR 1.17(e) has been timely paid, the finality of the previous Office action

has been withdrawn pursuant to 37 CFR 1.114. Applicant's submission filed on

December 13, 2007 has been entered.

Remarks

2.   In response to the RCE filed on December 13, 2007, claims 1-21 and 25-34 are

pending in this application.

3.   The rejections made under 35 USC 101 are overcome by the amendments to the

claims.

4.   The rejections made under 35 USC 112 2nd are withdrawn.

5.   The objection to claim 22 is withdrawn.

Claim Rejections - 35 USC § 103

6.   The following is a quotation of 35 U.S.C. 103(a) which forms the basis for all

obviousness rejections set forth in this Office action:

> (a) A patent may not be obtained though the invention is not identically disclosed or described as set
> forth in section 102 of this title, if the differences between the subject matter sought to be patented and
> the prior art are such that the subject matter as a whole would have been obvious at the time the
> invention was made to a person having ordinary skill in the art to which said subject matter pertains.
> Patentability shall not be negatived by the manner in which the invention was made.

7.   Claims 1-13, 15, 17-19, 21, and 25-34 are rejected under 35 U.S.C. 103(a) as

being unpatentable over Borkar et al., "Automatic segmentation of text strings into

structured records" and in view of Ando et al., "Mostly-Unsupervised Statistical

Segmentation of Japanese Sequences".

As to claims 1 and 27, Borkar et al. disclose:

A process (see Abstract, pg. 1, line 1) and system (see Abstract, pg. 1, paragraph 2,

line 1; wherein DATAMOLD is a system of interrelated components used to segment

text) to evaluate an input string to segment said input string into component parts

comprising:

means for providing a state transition model (see Abstract, pg. 1, paragraph 2, line 1

        DATAMOLD) based on an existing collection of data records that includes

        probabilities to segment input strings into component parts which adjusts said

        probabilities to account for token placement in the input string (see pg. 7, section

        2.5.1, lines 16-21);

means for determining a most probable segmentation (see Abstract, pg. 1, paragraph 2,

line 1 DATAMOLD) of the input string by comparing an order of tokens that make

up the input string with a state transition model derived from the collection of data

records (see pg. 3, section 1.3.1, col. 2, lines 9-11; wherein the inner HMMs

corroborate each other's findings to pick the segmentation that is globally

optimal).

means for segmenting the input string into one or more component parts according to

the most probable segmentation (see page 4, col. 2, lines 6-9 and 37-38); and

means for storing the one or more component parts in a data base (see abstract, line 7).

However, Borkar et al. do not explicitly disclose:

wherein the existing collection of data records does not comprise manually segmented

training data.

Ando et al. disclose:

wherein the existing collection of data records does not comprise manually segmented

training data (see abstract, lines 5-9 and page 2, lines 26-30).

It would have been obvious to have modified the teachings of Borkar et al. by the

teachings of Ando et al. to provide a simple, efficient segmentation method thus

avoiding the costs of hand-segmenting (manually segmenting) training data (see Ando

et al., page 2, lines 26-30).

As to claims 2 and 28, Borkar et al., as modified, disclose:

wherein the state transition model has probabilities for multiple states of said model and

a most probable segmentation is determined based on a most probable token emission

path through different states of the state transition model from a beginning state to an

end state (see <u>Borkar et al.</u>, pg. 4, col. 1, line 3; wherein the HMM has multiple states

and col. 2, lines 6-9 –path having the highest probability).

As to claims 3 and 29, <u>Borkar et al.</u>, as modified, disclose:

means for maintaining a collection of records, wherein the collection of data records is

stored in a database relation and an order of attributes for the database relation as the

most probable segmentation is determined (see <u>Borkar et al.</u>, pg. 3, Fig. 1; wherein the

structured record is determined and produced).

As to claims 4 and 30, <u>Borkar et al.</u>, as modified, disclose:

wherein the input string is segmented into sub-components which correspond to

attributes of the database relation (see <u>Borkar et al.</u>, pg. 1, col. 2, section 1.1, lines 5-

18).

As to claims 5 and 31, <u>Borkar et al.</u>, as modified, disclose:

wherein the tokens are substrings of said input string (see <u>Borkar et al.</u>, pg. 6, section

2.4, lines 2-4).

As to claims 6 and 32, <u>Borkar et al.</u>, as modified, disclose:

wherein the input string is to be segmented into database attributes and wherein each

attribute has a state transition model based on the contents of the database relation

(see <u>Borkar et al.</u>, pg. 4, Fig. 2; wherein each attribute has a transition in the model).

As to claims 7 and 33, <u>Borkar et al.</u>, as modified, disclose:

wherein the state transition model has multiple states for a beginning, middle and

trailing position within an input string (see <u>Borkar et al.</u>, pg. 6, Fig. 6; wherein state "1" is

the beginning, state "2" is the middle and state "3" is the trailing position).

As to claims 8 and 34, <u>Borkar et al.</u>, as modified, disclose:

wherein the state transition model has probabilities for the states and a most probable

segmentation is determined based on a most probable token emission [state] path

through different states of the state transition model from a beginning state to an end

state (see <u>Borkar et al.</u>, pg. 6, Fig 6 and col. 2, paragraph 2, lines 1-4).

As to claim 9, <u>Borkar et al.</u>, as modified, disclose:

wherein input attribute order for records to be segmented is known in advance of

segmentation of an input string (see <u>Borkar et al.</u>, Abstract, pg. 1, paragraph 2, lines 3-

8).

As to claim 10, <u>Borkar et al.</u>, as modified, disclose:

wherein an attribute order is learned from a batch of records that are inserted into the

table (see <u>Borkar et al.</u>, Abstract, pg. 1, paragraph 2, lines 1-3).

As to claim 11, <u>Borkar et al.</u>, as modified, disclose:

wherein the state transition model has at least some states corresponding to base

tokens occurring in the reference relation (see <u>Borkar et al.</u>, Abstract, pg. 1, paragraph

2, lines 1-8; wherein the training examples and dictionary provide the basis for

acceptable and recognizable input and therefore some states would correspond to the

same structure/ examples or base tokens).

As to claim 12, <u>Borkar et al.</u>, as modified, disclose:

wherein the state transition model has class states corresponding to token patterns

within said reference relation (see <u>Borkar et al.</u>, pg. 3, col. 1, paragraph 3, lines 1-8).

As to claim 13, <u>Borkar et al.</u>, as modified, disclose:

wherein the state transition model includes states that account for missing, misordered

and inserted tokens within an attribute (see <u>Borkar et al.</u>, pgs. 3-4, section 2; wherein

data mold uses the example segmented records to output a model that when presented

with any unseen text segments it into one or more of its constituent elements).

As to claim 15, <u>Borkar et al.</u>, as modified, disclose:

A machine computer readable medium containing instructions to perform the

evaluat[ion] [of] an input string to segment said input string into component parts (see
Borkar et al., pg. 1, section 1.1, lines 5-6; wherein the tool is used during warehouse
construction which implies that the program instructions are being read from a medium
inserted in or stored on a machine).

As to claim 17, Borkar et al. disclose:

    a) a database management system to store records organized into relations
       wherein data records within a relation are organized into a number of
       attributes (see page 1, Abstract, line 7 – corporate database);

    b) a model building component that builds a number of attribute recognition
       models based on an existing relation of data records, wherein one or more
       of said attribute recognition models includes probabilities for segmenting
       input strings into component arts which adjusts said probabilities to
       account for erroneous entries within an input string (see page 1, Abstract,
       lines 13-14; wherein DATAMOLD comprises a model building component
       because its built on HMM; and, (see pg. 7, section 2.5.1, lines 16-21
       accounting for invalid paths); and

    c) a segmenting component that receives an input string and determines a most
       probable record segmentation by evaluating transition probabilities of
       states within the attribute recognition models built by the model building
       component (see page 2, section 1.3, lines 1-3; wherein DATAMOLD
       comprises a segmenting component).

However, Borkar et al. do not explicitly disclose:

wherein the existing collection of data records does not comprise manually segmented training data.

Ando et al. disclose:

wherein the existing collection of data records does not comprise manually segmented training data (see abstract, lines 5-9 and page 2, lines 26-30).

It would have been obvious to have modified the teachings of Borkar et al. by the teachings of Ando et al. to provide a simple, efficient segmentation method thus avoiding the costs of hand-segmenting (manually segmenting) training data (see Ando et al., page 2, lines 26-30).

As to claim 18, Borkar et al., as modified, disclose:

wherein the segmenting component receives a batch of evaluation strings and determines an attribute order of strings in said batch and thereafter assumes the input string has tokens in the same attribute order as the evaluation strings (see Borkar et al., Abstract, pg. 1, paragraph 2, lines 3-8; wherein the training examples are the batch of strings that provide a basis for the structure of strings).

As to claim 19, Borkar et al., as modified, disclose:

wherein the segmenting component evaluates the tokens in an order in which they are contained in the input string and considers state transitions from multiple attribute recognition models to find a maximum probability for the state of a token to provide a

maximum probability for each token in said input string (see Borkar et al., pg. 4, section 2.1; wherein the segmenting component considers transitions from the multiple attribute states to find the maximum probability).

As to claim 21, Borkar et al., as modified, disclose:

wherein the model building component defines a start and end state for each model and accommodates missing attributes by assigning a probability for a transition from the start to the end state (see Borkar et al., pg. 6, Fig. 6).

As to claim 25, Borkar et al. disclose:

A process of segmenting a string input record into a sequence of attributes for inclusion into a database table comprising: wherein determining a most probable segmentation of the input string comprises:

considering a first token in a string input record and determining a maximum state

      probability for said token based on state transition models for multiple data table

      attributes (see pg. 4, section 2.1; wherein the segmenting component considers

      transitions from the multiple attribute states to find the maximum probability); and

considering in turn subsequent tokens in the string input record and determining

      maximum state probabilities for said subsequent tokens from a previous token

      state until all tokens are considered (see pg. 4, section 2.1; wherein the

      segmenting component considers transitions from the multiple attribute states to

      find the maximum probability); and

wherein segmenting the input string comprises segmenting the input string record by

> assigning the tokens of the string to attribute states of the state transition models
>
> corresponding to said maximum state probabilities
>
> (see pg. 4, Fig. 2, wherein the model displays attributes represented by states
>
> and section 2.1; wherein the segmenting component considers transitions from
>
> the multiple attribute states to find the maximum probability .

However, Borkar et al. do not explicitly disclose:

wherein the state transition models are based on an existing collection of data records

that does not comprise manually segmented training data.

Ando et al. disclose:

wherein the state transition models are based on an existing collection of data records

(sequences) that does not comprise manually segmented training data (see abstract,

lines 5-9 and page 2, lines 26-30).

> It would have been obvious to have modified the teachings of Borkar et al. by the

teachings of Ando et al. to provide a simple, efficient segmentation method thus

avoiding the costs of hand-segmenting (manually segmenting) training data (see Ando

et al., page 2, lines 26-30).

As to claim 26, Borkar et al., as modified, disclose:

additionally comprising determining an attribute order for a batch of string input records

and using the order to limit the possible state probabilities when evaluating tokens in an

input string (see <u>Borkar et al.</u>, Abstract, pg. 1, paragraph 2, lines 1-3; wherein the

structure and order] is learned from the training examples).

8.   Claims 14 and 20are rejected under 35 U.S.C. 103(a) as being unpatentable

over <u>Borkar et al.</u>; "Automatic segmentation of text strings into structured records", in

view of <u>Ando et al.</u>, "Mostly-Unsupervised Statistical Segmentation of Japanese

Sequences", and further in view of <u>Reed</u> (U.S. Pat. No. 5, 095, 432).

As to claim 14, <u>Borkar et al.</u> and <u>Ando et al.</u>, do not explicitly disclose:

wherein the state transition model has a beginning, a middle and a trailing state

topology and the process of accounting for misordered and inserted tokens is performed

by copying states from one of said beginning, middle or trailing states into another of

said beginning, middle or trailing states.

However, <u>Reed</u> discloses:

wherein the state transition model has a beginning, a middle and a trailing state

topology and the process of accounting for misordered and inserted tokens is performed

by copying states from one of said beginning, middle or trailing states into another of

said beginning, middle or trailing states (see col. 5, lines 1).

        It would have been obvious, at the time of the invention, having the teachings of

<u>Borkar et al.</u>, <u>Ando et al.</u>, and <u>Reed</u> before him/her, to combine the steps as disclosed

by <u>Borkar et al.</u> and <u>Ando et al.</u> with the feature as disclosed by <u>Reed</u> to enable

grammar developers to use the familiar PSG formalism to compile their grammars into

RVG for more efficient execution (see <u>Reed</u>, col. 2, lines 54-57).

As to claim 20, <u>Borkar et al.</u> and <u>Ando et al.</u>, do not explicitly disclose:

wherein the model building component assigns states for each attribute for a beginning,

middle and trailing token position (see pg. 4, Fig. 2; wherein the states are assigned to

each attribute and pg. 6, Fig. 6; wherein states are assigned for first (beginning state),

second (middle state), third (trailing state))

However, <u>Borkar et al.</u> does not explicitly disclose:

wherein the model building component relaxes token acceptance by the model by

copying states among said beginning, middle and trailing token positions.

<u>Reed</u> discloses:

wherein the model building component relaxes token acceptance by the model by

copying states among said beginning, middle and trailing token positions (see col. 5,

lines 1; wherein states in the transition model are copied).

It would have been obvious, at the time of the invention, having the teachings of

<u>Borkar et al.</u>, <u>Ando et al.</u>, and <u>Reed</u> before him/her, to combine the steps as disclosed

by <u>Borkar et al.</u> and <u>Ando et al.</u> with the feature as disclosed by <u>Reed</u> to enable

grammar developers to use the familiar PSG formalism to compile their grammars into

RVG for more efficient execution (see <u>Reed</u>, col. 2, lines 54-57).

9.   Claim 16 is rejected under 35 U.S.C. 103(a) as being unpatentable over <u>Borkar</u>
<u>et al.</u>; "Automatic segmentation of text strings into structured records" in view of <u>Ando</u>
<u>et al.</u>, "Mostly-Unsupervised Statistical Segmentation of Japanese Sequences", and
further in view of <u>Fairweather</u> (U.S. PG. Pub. No. 2006/0235811).

As to claim 16, <u>Borkar et al.</u> disclose:

providing a reference table of string records that are segmented into multiple substrings

         corresponding to database attributes (see Abstract, p. 1, paragraph 2, lines 1-3);

breaking the input record into a sequence of tokens, and determining a most probable

         segmentation of the input record by comparing the tokens of the input record with

         state models derived for attributes from the reference table (see pg. 3, section

         1.3.1, col. 2, lines 9-11; wherein the inner HMMs corroborate each other's

         findings to pick the segmentation that is globally optimal).

However, <u>Borkar et al.</u> does not explicitly disclose:

wherein the reference table of string records does not comprise manually segmented

         training data.

analyzing the substrings within an attribute to provide a state model that assumes a

         beginning, a middle and a trailing token topology for said attribute said topology

         including a null token for an empty attribute component;

<u>Ando et al.</u> disclose:

wherein the reference table of string records (sequences) does not comprise manually

         segmented training data. (see abstract, lines 5-9 and page 2, lines 26-30).

It would have been obvious to have modified the teachings of Borkar et al. by the

teachings of Ando et al. to provide a simple, efficient segmentation method thus

avoiding the costs of hand-segmenting (manually segmenting) training data (see Ando

et al., page 2, lines 26-30).

However, Borkar et al. and Ando et al. does not explicitly disclose:

analyzing the substrings within an attribute to provide a state model that assumes a

     beginning, a middle and a trailing token topology for said attribute said topology

     including a null token for an empty attribute component

Fairweather discloses:

analyzing the substrings within an attribute to provide a state model that assumes a

     beginning, a middle and a trailing token topology for said attribute said topology

     including a null token for an empty attribute component (see Fairweather,

     paragraph [0406], lines 8-9; wherein a the null pointer is returned because the

     token is null);

It would have been obvious, at the time of the invention, having the teachings of

Borkar et al., Ando et al., and Fairweather before him/her, to combine the steps as

disclosed by Borkar et al. and Ando et al. with the feature as disclosed by Fairweather

to provide a system in which the content of the data itself actually determines the order

of execution of statements in the mining language and automatically keeps track of the

current state (see Fairweather, paragraph [0004], lines 7-10).

Response to Arguments

10.   Applicant's arguments filed December 13, 2007 have been fully considered but
they are not persuasive.

Applicant's arguments that the cited reference do not disclose, "providing a state
transition model" and "adjusts said probabilities to account for token placement in the
input string " is acknowledged but are not considered to be persuasive.

Applicant argues that <u>Borkar et al.</u> does not disclose "providing a state transition
model" and "adjusts said probabilities to account for token placement in the input
string". The examiner disagrees in that <u>Borkar et al</u> clearly discloses a Hidden Markov
model or HMM which is an Finite State Automaton or FSA see Abstract, pg. 1,
paragraph 2, line 1 DATAMOLD). <u>Borkar et al</u> also clearly discloses changing the
equation so that instead of taking the max over all the states, they disallow the states
that are invalid (see pg. 7, section 2.5.1, lines 16-21).

Applicant's arguments that the cited reference do not disclose, "wherein the
existing collection of data records does not comprise manually segmented training data"
is acknowledged but are not considered to be persuasive.

Applicant argues that <u>Ando et al.</u> does not disclose "wherein the existing
collection of data records does not comprise manually segmented training data". The
examiner disagrees in that <u>Ando et al</u> clearly disclose using unsegmented training data
(see <u>Ando et al</u> abstract, lines 5-9).

Applicant's arguments that the cited reference do not disclose, "providing a reference table of string records that are segmented into multiple substrings corresponding to database attributes" is acknowledged but are not considered to be persuasive.

Applicant continues to argue that Ando et al. does not disclose "providing a reference table of string records that are segmented into multiple substrings corresponding to database attributes". Ando et al discloses an annotation scheme which, according to the examiner's interpretation, is the equivalent of applicant's reference table (see Abstract, p. 1, paragraph 2, lines 1-3).

Applicant's arguments that the cited reference do not disclose, "analyzing the substrings within an attribute to provide a state model that assumes a beginning, a middle and a trailing token topology for said attribute" is acknowledged but are not considered to be persuasive.

Lastly, applicant argues that Fairweather does not disclose "analyzing the substrings within an attribute to provide a state model that assumes a beginning, a middle and a trailing token topology for said attribute". The examiner disagrees in that Fairweather. discloses analyzing the substring and returning a the null pointer is returned because the token is null (see paragraph [0406], lines 8-9).

Conclusion

11.   Any inquiry concerning this communication or earlier communications from the examiner should be directed to Johnese Johnson whose telephone number is 571-270-1097. The examiner can normally be reached on 4/5/9.

If attempts to reach the examiner by telephone are unsuccessful, the examiner's supervisor, Hosain Alam can be reached on 571-272-3978. The fax phone number for the organization where this application or proceeding is assigned is 571-273-8300.

Information regarding the status of an application may be obtained from the Patent Application Information Retrieval (PAIR) system. Status information for published applications may be obtained from either Private PAIR or Public PAIR. Status information for unpublished applications is available through Private PAIR only. For more information about the PAIR system, see http://pair-direct.uspto.gov. Should you have questions on access to the Private PAIR system, contact the Electronic Business Center (EBC) at 866-217-9197 (toll-free). If you would like assistance from a USPTO Customer Service Representative or access to the automated information system, call 800-786-9199 (IN USA OR CANADA) or 571-272-1000.

17 February 2008
JJ

/Hosain T Alam/

Supervisory Patent Examiner, Art Unit 2166